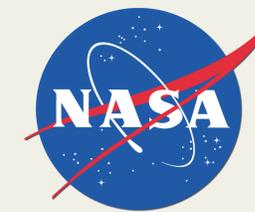




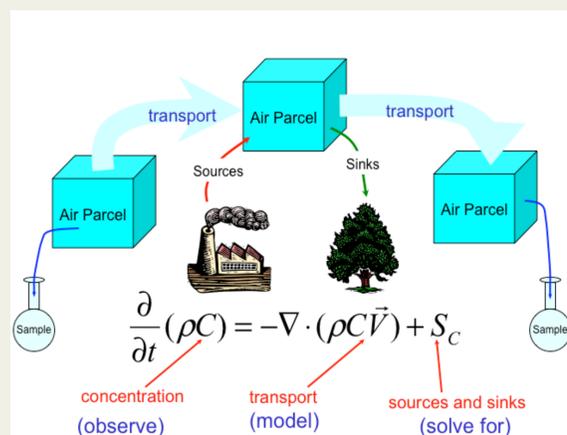
# Regional Carbon Budgets by Fusion of Multiple NASA Data Products



Scott Denning, Andrew Schuh, Katherine Haynes, and Ian Baker

CMMAP, Dept. of Atmospheric Science, Colorado State University, USA

## “Inverse Modeling” of Atmospheric CO<sub>2</sub>

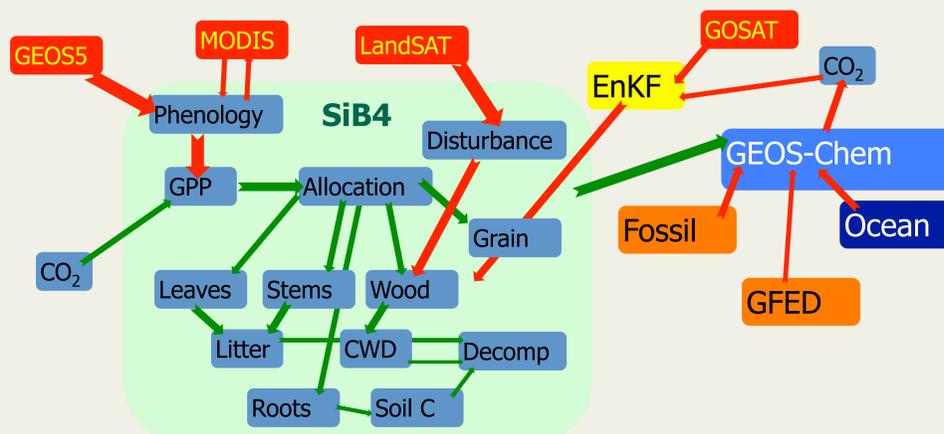


Seems simple enough, but:

- CO<sub>2</sub> “flux” is really just a **tiny difference (NEE)** between two much larger fluxes (GPP & Resp)
- Component fluxes are governed by **different BGC** processes, so will respond differently to future change
- NEE has **no “memory:”** knowing its value at one time is no help in estimating it later
- Direct estimation of NEE is therefore **poorly constrained**, even by satellite CO<sub>2</sub> observations

We seek improved constraint by instead **estimating long-lived state variables that control GPP and Resp**

## Forward Modeling and Data Assimilation



### Preliminary Test Optimizations

$$F_T(x, y, t) = (1 + \beta_{FF}(x, y))FF(x, y, t) + (1 + \beta_{Fire}(x, y))Fire(x, y, t) + (1 + \beta_{RESP}(x, y))RESP(x, y, t) - (1 + \beta_{GPP}(x, y))GPP(x, y, t) + (1 + \beta_{Ocean}(x, y))Ocean(x, y, t)$$

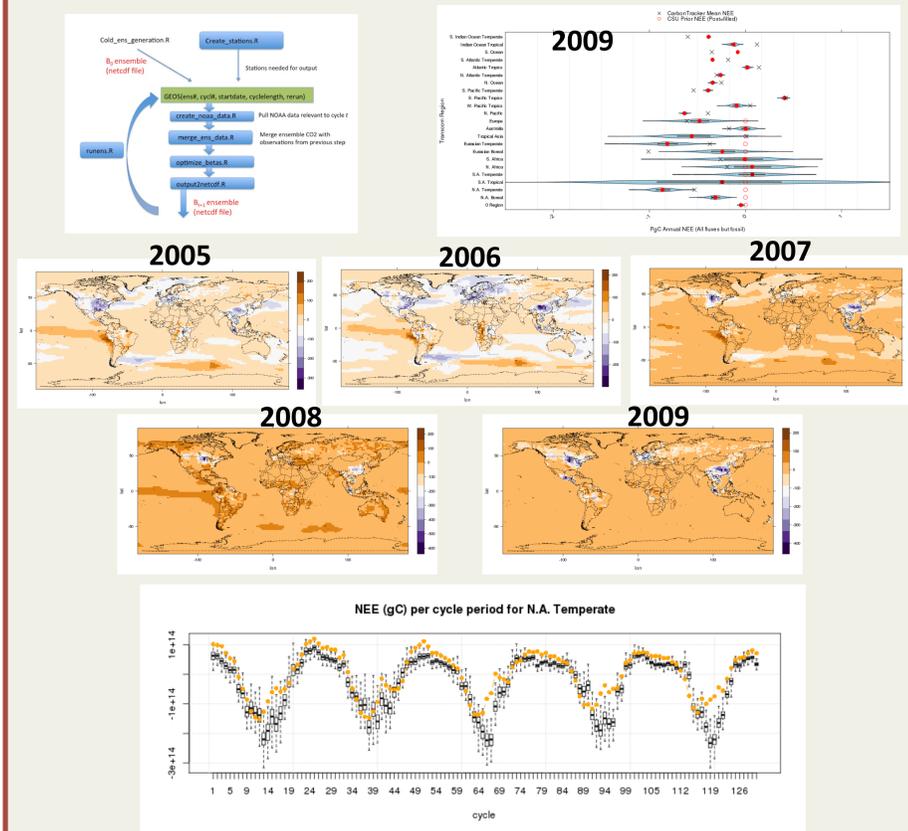
### Prototype of Future Optimization

GPP estimation from **photosynthetic capacity** (fPAR \* fractional coverage \* Vmax)

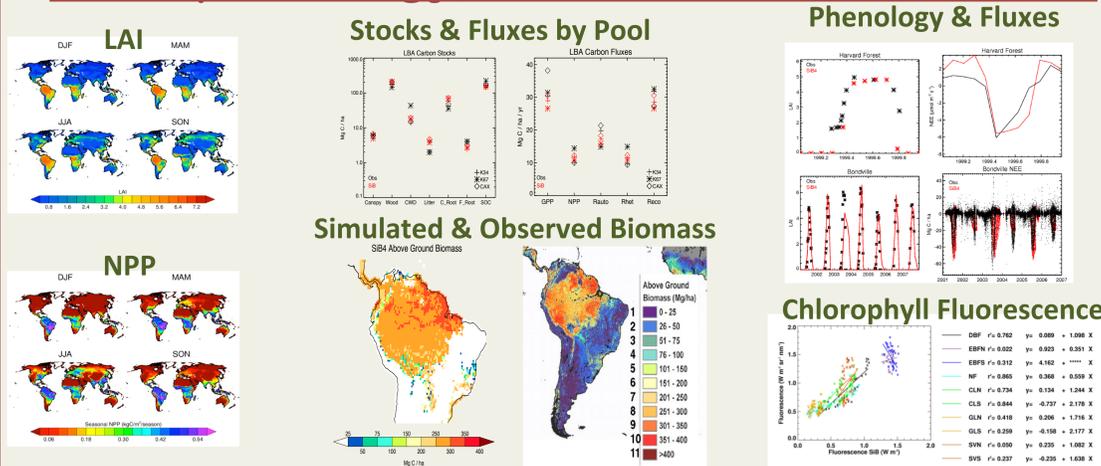
Resp estimation from **decomposing biomass** (litter + intermediate soil carbon + CWD)

## Preliminary Optimization from Atmospheric CO<sub>2</sub>

Using NOAA in-situ data



## SiB4: phenology, biomass, & fluorescence



## Using GOSAT Data

